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Article

Carbon Emissions Reduction and Corporate Financial Performance: The Influence of Country-Level Characteristics

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Abstract: Using a cross-country dataset covering 9265 observations on 1785 firms representing 53 countries over the period 2004–2019, this study investigates the relation between carbon emissions reduction and corporate financial performance (CFP). We perform OLS regressions with fixed effects. We found that carbon emissions reduction increases the return on assets, the return on equity, and the return on sales, whereas it has no effect on the Tobin's Q and the current ratio. The positive relationship with the return on assets is stronger for firms with a higher responsibility score. We study country characteristics by modeling GDP growth, overall emissions within a country, and the presence of carbon emissions legislation. Our results indicate that the overall carbon emissions of a country and the presence of carbon emissions legislation are related to both corporate carbon emissions reduction and CFP. Moderating effects of the country's overall emissions and the presence of carbon emissions legislation do not affect the relationship between carbon emissions reduction and CFP. Despite the further understanding gained, the issue of whether it “pays to be green” can still not be resolved well.

Keywords: carbon emissions reduction; corporate financial performance; country-level characteristics

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1. Introduction

Climate change is one of the most threatening and complex challenges the world has ever faced [1]. The main cause of climate change is carbon emissions that are released into the air [2]. The challenge of climate change is gaining ever more attention. Countries and organizations they constitute, such as the United Nations and the European Union, are all around the world aiming to reduce overall carbon emissions. For instance, carbon emissions reduction is one of the United Nations Sustainable development goals to reduce the amount of greenhouse gasses by 45% by 2030 [3]. The European Union aims to reduce greenhouse emissions by 55% by 2030 [4]. These goals show the importance of the reduction of the overall greenhouse gasses emission worldwide. Legislators attempt to encourage or enforce firms to reduce their overall emissions, but it is unclear how firms respond to this threat of climate change.

Moreover, existing studies still leave the debate open about whether it pays to be green. On the one side [5], they claim that more polluting firms have a better corporate financial performance (hereafter: CFP). The main argument is that being or becoming green requires investments and it is not certain that these will be earned back. Delmas et al. [6] show that becoming green reduces the short-term profitability of firms but pays off in the long term. This short-term decrease in profitability may affect the decisions making of managers due to short-term performance targets. In addition, these arguments can provide an answer to the issue of why firms respond to the climate change issue slowly [7]. On the other hand, researchers have found that firms with lower emissions have a better

CFP compared to more polluting firms [8]. Fujii et al. [9] show that firms with lower emissions had better profitability and a higher capital turnover. Fernández-Cuesta et al. [10] found that firms with a better carbon performance were able to obtain more long-term financial debt to finance their environmental investments. Gallego-Alvarez et al. [11] found that during times of economic crisis, the synergy between environmental and financial performance is higher, indicating that firms must invest in sustainable projects to enhance their relationship with stakeholders even during crisis times.

Our study contributes by exploring the effect of carbon emissions reduction on the financial performance of firms by using the following financial performance indicators: return on assets (hereafter: ROA), return on equity (hereafter: ROE), return on sales (hereafter: ROS), Tobin's Q, and (as a new element) the current ratio (hereafter: CR). Gallego-Alvarez et al. [12] suggest that future research into the relationship between CFP and carbon emissions reduction should include the effect of country characteristics in a large cross-country sample. Their suggestion formed the foundation for this research. This study will look at an extraordinarily large cross-country sample including 1785 firms covering 53 countries over the period 2004–2019, by using financial and environmental firm-level data from the Thomson Reuters Eikon Database. Moreover, we investigate the effect of overall carbon emissions, economic development, and the presence of carbon emissions legislation within the home country of a firm, by using country-level data from the World Bank Database. We test this by including countries' overall emissions and a dummy that indicates the presence of carbon emissions legislation within a country as a moderating variable. The inclusion of the effect of carbon emissions legislation connects to a suggestion made by Lewandowski [7] that future research should focus on the incentives firms have to reduce their emissions. We control for overall growth in GDP by including this country characteristic as well [13].

2. Literature Review and Hypotheses Development

This section covers the existing literature used to provide background for this study. The first part focuses on the relationship between carbon emissions reduction and CFP; the second part includes the moderating effect of country characteristics on the relationship between carbon emissions reduction and CFP, and the third part includes the firm's level of responsibility as moderating effect on the relationship between carbon emissions reduction and CFP.

2.1. Carbon Emissions Reduction and CFP

The relation between carbon emissions and CFP has been studied before. Busch et al. [5] found that more polluting firms performed better financially compared to firms with lower emissions. However, Busch and Lewandowski [14] found evidence that firms with lower emissions performed better financially. In addition, studies on the relation between carbon mitigation and CFP also showed mixed results. Gallego-Alvarez et al. [12] found that carbon emissions reduction improves the CFP in terms of return on equity and return on assets. Lewandowski [7] found that carbon emissions reduction leads to an increase in the return on sales but a decrease in CFP measured by Tobin's Q. Kim et al. [15] showed that carbon mitigation reduces the costs of capital, this reduction can be used to overcome costs that come along with emission reduction. Delmas et al. [6] found that the short-term profitability of firms is affected by carbon emissions reduction, but also showed that Tobin's Q improves due to the emission reduction, suggesting that investors see the potential long-term value of the mitigation of the emissions. The goal of this study is to understand the relation between carbon emissions reduction and CFP and to assess the effect of country characteristics on this relationship, as suggested by Gallego-Alvarez et al. [12]. The main research question can be stated briefly as follows: "*Does carbon emissions reduction impact the CFP of firms?*"

In addition, the measurement of CFP differs among the different studies. Lewandowski [7] uses the ROA, ROE, ROS, ROIC, and Tobin's Q. Delmas et al. [6] use the ROA

and Tobin's Q. Gallego-Alvarez et al. [12] use the ROA and the ROE. Most of the researchers, therefore, mainly focus on accounting profitability measures. To cover the effect of the market value of firms, the Tobin's Q is included since the Tobin's Q ratio reflects the expectations of the stock market on the future profitability and growth of firms [16]. Based on the mixed results, we formulate the main hypothesis without direction, as follows:

Hypothesis 1 (H1). *Carbon emissions reduction influences the CFP of firms.*

2.2. Carbon Emissions Reduction, CFP, and Country Characteristics

We investigate the effect of two country characteristics. First, we look at the effect of the overall carbon emissions of a country to see whether the effect of carbon emissions reduction is different within more polluting countries. Alonso-Martinez et al. [17] found that the overall carbon emissions of a country are related to the environmental, social, and governance (ESG) responsibility of firms. They suggest that higher levels of pollution raise the awareness towards environmental sustainability within a country. Jiménez-Parra et al. [18] claim that the concerns around air pollution lead to greater environmental demands by stakeholders. The second hypothesis deals with pollution within a country and reads as follows:

Hypothesis 2 (H2). *The relationship between carbon emissions reduction and CFP is influenced by the overall carbon emissions within a country.*

Firms have multiple incentives to reduce their overall emissions, which are driven by the pressures of different stakeholders thriving on the importance of climate awareness [19]. The main stakeholder used in this study is the legal authorities that implement regulations to achieve overall greenhouse gases reductions. These days, more and more major organizations such as the United Nations, the European Union, and a wide variety of countries set climate goals to reduce their overall emissions. To achieve these goals, they introduced multiple market-based mechanisms that can be distinguished into two groups: quantity control and price control [20]. Cap and trade emissions trading schemes are the most important as to quantity control, with 27 trading schemes being implemented worldwide [21]. In a cap and trade system, companies with higher carbon emissions have to lower their emissions or have to buy additional allowances, whereas companies with a lower level of emissions are able to sell their surplus of allowances [22]. The price control mechanism focuses on the taxation of carbon emissions. Both systems are implemented in a wide variety of countries. A meta-analysis by Galama and Scholtens [8] indicates that the positive relationship between CFP and CEP is higher for countries that have more stringent carbon regulations. New to the existing body of work is that we include the presence of a market-based mechanism within a country as a moderating variable.

The impact of cap and trade systems on CFP has been previously researched; however, the results are mixed. Brouwers et al. [23] found that only firms that are not able to pass their costs to their customers are negatively impacted. Griffin [24] shows that the net income of Californian firms was negatively impacted by the AB32, the Californian cap, and trade system. Moreover, Delmas et al. [6] found that the short-term profitability of firms is negatively impacted by carbon emissions reduction, which is the main goal of these programs. Marin et al. [25] show that the EU Emission Trading Scheme (henceforth: EU ETS) has a positive impact on the turnover, investment intensity, and labor productivity of firms. De Giovanni and Vinzi [26] found no relationship between the impact of the EU ETS and CFP. Jong et al. [27] show that shareholders regard the EU ETS as value-relevant, which is in line with the positive relation Delmas et al. [6] found between Tobin's Q and carbon emissions reduction. Taxation is the most important quality control mechanism, being fully implemented in 25 countries. Luo and Tang [28] researched the impact of the implementation of carbon tax and concluded that shareholders value is negatively associated with the implementation of carbon tax regulations, which can be caused by an

increase in competition of firms in countries without carbon taxation [29]. Their results also show that the carbon emissions of the firms within their sample were negatively correlated to their abnormal returns.

To examine the effect of regulatory forces to enforce carbon emissions reduction, the third hypothesis focuses on the effects of carbon emissions legislation within a country. Galama and Scholtens [8] indicate that the positive relation between CFP and CEP is stronger in countries with more stringent climate policies. To investigate whether such a claim stands, we include a variable measuring the presence of carbon emissions legislation. Moreover, we look into two different mechanisms: carbon emissions trading schemes and carbon taxation.

Hypothesis 3 (H3). *The relationship between carbon emissions reduction and CFP is influenced by the presence of carbon taxation or emission trading schemes within a country.*

2.3 Carbon Emissions Reduction, CFP, and Responsibility

Russo and Fouts [30] argue that firms that act responsibly to reduce their environmental footprint can create a competitive advantage. Beck et al. [31] argue that firms that engage more in responsible activities have a better CFP. Sariannidis et al. [32] researched the effect of carbon emissions on the CFP of responsible firms. Their main finding is that responsible firms react negatively to a global increase in overall emissions. Jiménez-Parra et al. [18] found that the presence of more responsible firms reduces the overall emissions within a country. To investigate whether the relation between carbon emissions reduction and CFP is different for responsible firms, we create the fourth hypothesis:

Hypothesis 4 (H4). *The relationship between carbon emissions reduction and CFP is influenced by a firm's level of responsibility.*

3. Results

3.1. Data

The existing body of work [5–7,12] provides insights into carbon emissions reduction effects; however, these studies contain limitations. Their first main limitation refers to the relatively small sample size and period. Gallego-Alvarez [12] only used 257 observations. Delmas et al. [6] have 1095 (ROA) and 880 observations (Tobin's Q). Busch et al. [5] expand these amounts to 2884 and 2896, respectively. Our study overcomes this limitation by extending the sample period to 2000–2020 and using a large cross-country sample of 9265 observations.

Another limitation of the existing literature is endogeneity. To encounter this limitation, we will follow the methodology of Busch et al. [5] and lag the dependent variable. In addition, we include country characteristics and multiple control variables to reduce the threat of endogeneity. Lewandowski [7] points out that the carbon intensity of firms might be correlated to the CFP indicators. Delmas et al. [6] also describe the limitation of the current measuring methods. To address this issue, we will not only include profitability ratios, but also solvency and liquidity ratios.

The existing research uses multiple databases to measure the carbon emissions of the companies. Delmas et al. [6] describe the quality of their emission data as a limitation, since they use data from Trucost that also contains estimated emissions. They suggest future research only uses the reported carbon emissions. We follow Lewandowski [7] by using the ASSET4 Database. The carbon emission data can be separated into three scopes: scope 1 covers all the direct emissions from production processes on-site, scope 2 covers all indirect emissions from the usage of purchased energy, and scope 3 includes all other emission sources [6,32]. This research will follow Lewandowski [7] by only including the sum of the emission data for scope 1 and scope 2. Lewandowski argues that the quality of

the data for scope 3 is not very high and the carbon emissions within scope 3 are not impacted by regulations and, therefore, are not a significant cost driver.

This research uses a large global sample covering 53 countries. The observations are gathered over the period 2004–2019. The final sample contains 9265 observations covering 1785 firms. The financial and carbon emission data is drawn from the Thomson Reuters Eikon Database. This database contains data for 50,000 active firms covering 125 markets [33]. The emission data is part of the ASSET4 Database, which contains large amounts of responsibility data. The country-level data is gathered from the World Bank Database. Observations with negative sales and assets are eliminated. The two types of data are merged and winsorized at the 1 and 99 percent levels to eliminate outlier effects. In finance research, it is common that financial firms are excluded. Lewandowski [7] and Delmas et al. [6] argue that financial firms have low values for the ROA and carbon emissions though. A robustness check shows that the presence of financial firms does not affect the results.

3.2. Variables

Appendix A provides an overview of the variables that are used within this research. The independent variable used in this research is *DELTA*CO₂, which represents the change in carbon intensity over a year. It is calculated as the annual change in emissions scaled by the annual net sales to account for changes caused by changes in the scale of operations. The enlargement or reduction of company operations the total carbon emissions are divided by the total sales within year *t*. To reduce the effect of endogeneity, the independent variable *DELTA*CO₂ is lagged with *t*–1. Existing literature [7,34] uses the squared of the variation in carbon intensity to account for a curvilinear relationship between carbon performance and CFP. We name it *DELTA*CO₂².

The dependent variable within this study is *CFP*, which represents the corporate CFP of firms. In this research, we use multiple variables that measure the CFP: ROA, ROS, ROE, and Tobin's Q. In addition to the existing literature, we also include the current ratio, which measures the firm's ability to repay short-term debts [35]. This research uses multiple profitability ratios to measure the CFP. To reduce the impact of multicollinearity, we follow the methodology of Gallego-Alvarez et al. [12]. They calculate the return on assets with the usage of the operating income instead of the net income, which is already used in the return on sales and return on equity. Because of the availability and quality of the actual data, we follow the method by Chung and Pruitt [36] to calculate Tobin's Q. It is the sum of the market capitalization, current liabilities net of the current assets, and the book value of the long-term debts divided by the total assets of a firm. Next to the profitability and stock market performance indicators, we include a liquidity ratio. The current ratio, *CR*, reflects the firm's ability to repay short-term liabilities. Durrah et al. [35] argue that the firm's liquidity influences the CFP, since liquidity problems can affect the firm's level of operations and a higher level of liquidity can benefit the firm's level of operations due to more available funds for, e.g., investments. *CR* is calculated by dividing the current assets by the current liabilities.

This research includes several moderating variables. The first moderating variable is *CO2EMISSION*, which measures the average overall carbon emissions of the home country of the firm as a ratio of a country's overall GDP. We took the average value over the last 20 years. We also compute the variable *LEGAL*, which represents a dummy variable that indicates the presence or participation in a trading emission scheme or carbon taxation program in a certain country or year. An overview of the presence of carbon emissions legislation by country is provided in Appendix B. The variable *LEGAL* provides insight into the reasoning of firms to reduce their emissions, whether this is forced by regulations or carried out voluntarily. We also compute the variable *ESG* that measures the responsibility level of a firm.

To reduce the impact of omitted variable bias, we compute a wide variety of firm and county-level control variables. The first variable we control for is *SIZE*, by using the logarithm of the total market capitalization. The second control variable is *LEVERAGE*, measured by dividing the long-term debts by the total assets. The third variable we control for is *CAPINT*, which represents the capital intensity of firms. It is calculated by dividing the total equity by the total sales. The fourth control variable is *DELTASALES*, which represents the ratio of the differences in sales between year *t* and year *t*−1. The fifth control variable, *CASHFLOW*, represents the firm's free cash flow at the end of year *t*. It is presented as a ratio of the annual sales of the firm. Finally, we control for the economic development of the country where the firm is registered. The variable *GDPGROWTH* represents the annual growth in GDP compared to year *t*−1. The GDP is defined as the sum of all gross value added by producers plus product taxes and minus any subsidies that are not yet included.

3.3. Methodology

We use OLS regression to estimate the effect of carbon emissions reduction on CFP and to test the main and alternative hypotheses. We include year, country, and industry fixed effects to control for differences caused by unobserved time-invariant characteristics. We cluster all the standard errors on the firm level. To reduce the effect of endogeneity, the main independent variable *DELTACO2* and the variable *DELTACO2*² are lagged with *t*−1 throughout all the models. CFP represents the financial performance of firm *i* in year *t*, and is measured by *ROA*, *ROE*, *ROS*, *TOBIN*, and *CR*.

The first model tests for the relation between carbon emissions reduction and CFP. The regression function used is the following:

$$\begin{aligned} \text{CFP}_{i,t} &= \beta_0 + \beta_1 \text{DELTACO2}_{i,t-1} + \beta_2 \text{DELTACO2}^2_{i,t-1} + \beta_3 \text{CONTROL VARIABLES} \\ &+ \text{fixed effects}_{i,c,t,u} + \varepsilon_t \text{CFP}_{i,t} \\ &= \beta_0 + \beta_1 \text{DELTACO2}_{i,t-1} + \beta_2 \text{DELTACO2}^2_{i,t-1} \\ &+ \beta_3 \text{CONTROL VARIABLES} + \text{fixed effects}_{i,c,t,u} + \varepsilon_t \end{aligned} \quad (1)$$

Model 2 introduces the three moderating variables to test the alternative hypotheses: *CO2EMISSION*, *LEGAL*, and *ESG*. We employ the following regression equation:

$$\begin{aligned} \text{CFP}_{i,t} &= \beta_0 + \beta_1 \text{DELTACO2}_{i,t-1} + \beta_2 \text{DELTACO2}^2_{i,t-1} + \beta_3 \text{CONTROL VARIABLES}_{i,t} \\ &+ \beta_4 \text{C_CO2}_{i,t} + \beta_4 \text{LEGAL}_{i,t} + \beta_4 \text{ESG}_{i,t} + \text{fixed effects}_{i,c,t,u} + \varepsilon_t \text{CFP}_{i,t} \\ &= \beta_0 + \beta_1 \text{DELTACO2}_{i,t-1} + \beta_2 \text{DELTACO2}^2_{i,t-1} \\ &+ \beta_3 \text{CONTROL VARIABLES}_{i,t} + \beta_4 \text{C_CO2}_{i,t} + \beta_4 \text{LEGAL}_{i,t} + \beta_4 \text{ESG}_{i,t} \\ &+ \text{fixed effects}_{i,c,t,u} + \varepsilon_t \end{aligned} \quad (2)$$

Model 3 tests include three interaction variables to measure the effect of a certain variable on the relation between carbon emissions reduction and CFP. The first of these is *DELTACO2 X CO2*, which measures the effect of the overall emissions in the country where firm *i* is registered. The second interaction variable is *DELTACO2 X LEGAL*, which indicates the effect of the presence of carbon emissions legislation in the country where firm *i* is registered. The third interaction variable, *DELTACO2 X ESG*, measures the effect of the company's responsibility score on the relation between carbon mitigation and CFP. We use the following regression equation:

$$\begin{aligned}
CFP_{i,t} &= \beta_0 + \beta_1 DELTACO2_{i,t-1} + \beta_2 DELTACO2^2_{i,t-1} + \beta_3 CONTROL\ VARIABLES_{i,t} \\
&+ \beta_4 DELTACO2 \times CO2_{c,t} + \beta_5 DELTACO2 \times LEGAL_{c,t} \\
&+ \beta_6 DELTACO2 \times ESG_{i,T} + \text{fixed effects}_{i,c,t,u} + \varepsilon_t CFP_{i,t} \\
&= \beta_0 + \beta_1 DELTACO2_{i,t-1} + \beta_2 DELTACO2^2_{i,t-1} \\
&+ \beta_3 CONTROL\ VARIABLES_{i,t} + \beta_4 DELTACO2 \times CO2_{c,t} \\
&+ \beta_5 DELTACO2 \times LEGAL_{c,t} + \beta_6 DELTACO2 \times ESG_{i,T} + \text{fixed effects}_{i,c,t,u} \\
&+ \varepsilon_t
\end{aligned} \tag{3}$$

4. Results

4.1. Descriptive Statistics

Table 1 provides descriptive statistics for the sample covering 9265 observations. The mean value for *DELTACO2* states that the firms within the sample are overall reducing their emissions. The negative median value for this coefficient tells that over 50% of the observations contain a negative value, which means that firms are reducing their overall emissions compared to their sales. The positive mean and median values for *ROA*, *ROE*, and *ROS* show that the firms within the sample are overall profitable during the period 2004–2019. The mean value for the coefficient *TOBIN* of 1.314 shows that on average the market value of the firms within the sample is higher than the book value of their assets. The mean value of the variable *CO2EMISSION* is 0.312. Countries on average emit 0.312 kg CO₂ per unit of GDP. The lower median value of 0.285 and the maximum value of 0.849 shows that the average emissions are driven by some specific countries. The mean value for the variable *LEGAL* of 0.752 indicates that on average 75.2% of the observations represent a firm that is registered in a country with carbon emissions legislation. The variable *ESG* measures the firm's level of responsibility on a scale from 1 to 100. The mean value of the sample is 55.619 and the median value is 55.847 suggesting that the data for this variable is symmetrically distributed.

Table 1. Descriptive statistics.

Variable	N	Mean	Median	STD	Min	Max
<i>DELTACO2</i>	9265	−0.00000770	−0.00000006	0.00010950	−0.00063240	0.00050330
<i>DELTACO2</i> ²	9265	0.00000002	0.00000001	0.00000014	0.00000000	0.00000127
<i>ROA</i>	9265	0.068	0.065	0.075	−0.238	0.285
<i>ROE</i>	9265	0.099	0.057	0.347	−0.705	2.936
<i>ROS</i>	9265	0.108	0.109	0.322	−1.636	1.623
<i>TOBIN</i>	9265	1.314	0.897	1.485	−0.091	10.333
<i>CR</i>	9265	1.554	1.344	0.863	0.400	5.257
<i>SIZE</i>	9265	23.805	23.491	2.492	18.123	29.975
<i>LEVERAGE</i>	9265	0.197	0.180	0.131	0.000	0.577
<i>CAPINT</i>	9265	2.537	1.366	6.478	0.344	59.448
<i>DELTASALES</i>	9265	0.049	0.035	0.163	−0.425	0.783
<i>CASHFLOW</i>	9265	0.209	0.114	0.610	−0.381	5.560
<i>GDPGROWTH</i>	9265	0.017	0.019	0.020	−0.091	0.252
<i>CO2EMISSION</i>	9265	0.312	0.285	0.151	0.109	0.849
<i>LEGAL</i>	9265	0.752	1.000	0.432	0.000	1.000
<i>ESG</i>	9265	55.619	55.847	15.344	19.753	87.065

Descriptive statistics for the entire sample, covering 9265 observations representing 53 countries. The independent variables are *DELTACO2* and *DELTACO2*², which are both lagged with *t*−1. The dependent variables are *ROA*, *ROE*, *ROS*, *TOBIN*, and *CR*. The control variables are *SIZE*, *LEVERAGE*, *CAPINT*, *DELTASALES*, *CASHFLOW*, and *GDPGROWTH*. The moderating variables covering the alternative hypotheses are *CO2EMISSION*, *LEGAL*, and *ESG*. The firm-level data is collected from the Thomson Reuters Eikon Database and the country-level data is obtained with the usage of the World

Bank Database. The sample covers observations over the period 2004–2019. All firm-level data is winsorized at the 1 and 99 percentile.

4.2. Pearson Correlation Matrix

Table 2 provides the Pearson correlation matrix. Values above 0.8 are considered as a strong correlation. The results provided in Table 2 show two correlations above the threshold of 0.8. The first strong correlation is between the variables *ROS* and *CASHFLOW*, with a value of 0.878, the second strong correlation is between the variables *CAPINT* and *CASHFLOW* with a value of 0.928. To see whether this high correlation causes multicollinearity, we conduct a variance inflation factor (hereafter: VIF) analysis on our regression models. For this test, we look at the regression model that includes the moderating variables but does not include the interaction variables. There is no single VIF value above 10, which is regarded as a rule of thumb hurdle on multicollinearity.

Table 2. Pearson correlation matrix.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)
(1) <i>DELTA</i> <i>CO</i> <i>2</i>	1														
(2) <i>ROA</i>	−0.053	1													
(3) <i>ROE</i>	−0.016	0.528	1												
(4) <i>ROS</i>	−0.006	0.355	0.229	1											
(5) <i>TOBIN</i>	−0.023	0.369	0.185	0.162	1										
(6) <i>CR</i>	0.005	0.123	−0.007	0.027	0.057	1									
(7) <i>SIZE</i>	0.016	0.163	0.095	0.180	0.139	0.036	1								
(8) <i>LEVERAGE</i>	0.002	−0.039	0.037	−0.040	0.076	−0.183	−0.161	1							
(9) <i>CAPINT</i>	0.026	0.012	−0.005	0.753	0.051	−0.018	0.087	0.049	1						
(10) <i>DELTA</i> <i>SALE</i> <i>S</i>	−0.016	0.240	0.113	0.162	0.093	0.034	0.068	−0.027	0.068	1					
(11) <i>CASHFLOW</i>	0.021	0.198	0.103	0.878	0.117	−0.005	0.139	0.006	0.928	0.116	1				
(12) <i>GDPGROWT</i> <i>H</i>	0.029	0.061	0.043	0.193	0.113	0.016	0.022	0.020	0.214	0.047	0.222	1			
(13) <i>CO</i> <i>2</i> <i>EMISSIO</i> <i>N</i>	−0.027	0.002	−0.021	0.033	0.028	0.070	0.003	−0.009	0.040	0.047	0.039	0.130	1		
(14) <i>LEGAL</i>	0.020	0.014	0.026	−0.128	−0.021	−0.026	−0.216	0.083	−0.154	−0.074	−0.150	−0.144	−0.351	1	
(15) <i>ESG</i>	−0.009	0.104	0.067	0.064	0.083	−0.068	0.155	0.041	0.029	−0.011	0.049	−0.014	−0.086	0.049	1

The Pearson correlation matrix shows the correlation among the different variables. The coefficients are based on the full sample of 9265 observations. The matrix shows the correlation of the independent variable *DELTA**CO**2*, the dependent variables *ROA*, *ROE*, *ROS*, *TOBIN*, and *CR*, the control variables *SIZE*, *LEVERAGE*, *CAPINT*, *DELTA**SALES*, *CASHFLOW*, and *GDPGROWTH* and the moderating variables *CO**2**EMISSION*, *LEGAL*, and *ESG*. The variable definitions can be found in Table A1.

4.3. Regression Analysis *ROA*, *ROE*, and *ROS*

We now provide the results of the OLS regression on the relation between carbon emissions reduction and CFP as denoted by three variables: *ROA*, *ROE*, and *ROS*. Table 3 provides three different models. Model 1 only contains the dependent variable and the control variables, Model 2 introduces the moderating variables, and Model 3 introduces the created interaction variables for these moderating effects. The first dependent variable to measure CFP is *ROA*, which represents the return on assets. The regression results can be found in Models 1–3 of Table 3. Looking at the coefficient of *DELTA**CO**2* throughout

Models 1–2, the results show a strong significant negative relation between CFP and carbon emissions reduction ($\beta_1 = -31.7559$, $\beta_2 = -31.8168$, $p < 0.01$). The reduction in carbon emissions relates positively to a firm's return on assets. The value shown in the first Model ($\beta_1 = -31.7559$) shows that if a firm reduces its greenhouse emissions by 100%, its return on assets increases by 31.76% ceteris paribus. In Model 3, the coefficient remains negative but at a higher confidence interval ($\beta_2 = -66.2760$, $p < 0.1$). These findings are in line with Gallego-Alvarez et al. [12] who found a similar relation but which contradicts Delmas et al. [6] who found a negative relation between carbon emissions reduction and short-term profitability. The variable $DELTA_{CO2}^2$ shows negative but insignificant coefficients. These findings do not show a curvilinear relation between CFP in terms of return on assets and carbon emissions reduction. This contradicts the results of Lewandowski [7], who found evidence for a curvilinear relation between ROA and carbon mitigation.

Table 3. Regression analysis on ROA and ROE.

CFP Measure	ROA	ROA	ROA	ROE	ROE	ROE
Variable	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6
<i>DELTA_{CO2}</i>	-31.7559 *** [7.3145]	-31.8168 *** [7.3349]	-66.2760 * [34.4802]	-67.2307 ** [32.0351]	-67.1948 ** [32.1056]	-116.407 [125.0350]
<i>DELTA_{CO2}²</i>	-9020 [6243.3670]	-8880 [6245.1541]	-9580 [6263.9008]	-59,500** [28,500]	-59,500 ** [28,400]	-59,100 ** [27,900]
<i>SIZE</i>	0.0106 *** [0.0010]	0.0102 *** [0.0010]	0.0101 *** [0.0010]	0.0340 *** [0.0048]	0.0326 *** [0.0048]	0.0326 *** [0.0048]
<i>LEVERAGE</i>	-0.0217 ** [0.0110]	-0.0225 ** [0.0110]	-0.0225 ** [0.0110]	0.0916 [0.0585]	0.0896 [0.0585]	0.0897 [0.0585]
<i>CAPINT</i>	-0.0113 *** [0.0019]	-0.0112 *** [0.0019]	-0.0113 *** [0.0019]	-0.0279 *** [0.0059]	-0.0278 *** [0.0059]	-0.0278 *** [0.0059]
<i>DELTA_{SALES}</i>	0.0640 *** [0.0060]	0.0645 *** [0.0060]	0.0645 *** [0.0060]	0.1269 *** [0.0227]	0.1277 *** [0.0228]	0.1274 *** [0.0229]
<i>CASHFLOW</i>	0.1320 *** [0.0201]	0.1318 *** [0.0200]	0.1320 *** [0.0199]	0.3362 *** [0.0634]	0.3356 *** [0.0633]	0.3359 *** [0.0634]
<i>GDPGROWTH</i>	0.0063 [0.0513]	0.0036 [0.0511]	-0.0032 [0.0509]	-0.1490 [0.2117]	-0.1596 [0.2124]	-0.1616 [0.2118]
<i>CO2EMISSION</i>		0.0190 [0.0252]	0.0196 [0.0253]		0.1755 ** [0.0802]	0.1762 ** [0.0800]
<i>LEGAL</i>		0.0011 [0.0030]	0.0012 [0.0029]		0.0122 [0.0126]	0.0126 [0.0125]
<i>ESG</i>		0.0002 *** [0.0001]	0.0002 *** [0.0001]		0.0006 * [0.0003]	0.0006 * [0.0003]
<i>DELTA_{CO2}_X__{CO2}</i>			-56.6007 [46.3057]			39.6869 [180.2594]
<i>DELTA_{CO2}_X__{LEGAL}</i>			10.4208 [18.6720]			46.9072 [77.3637]
<i>DELTA_{CO2}_X__{ESG}</i>			0.8079 * [0.4263]			0.0477 [1.9778]
CONSTANT	-0.1729 *** [0.0267]	-0.1735 *** [0.0315]	-0.1741 *** [0.0316]	-0.7536 *** [0.1208]	-0.7841 *** [0.1339]	-0.7832 *** [0.1340]
FIXED EFFECTS	Yes	Yes	Yes	Yes	Yes	Yes
Adjusted R-squared	0.4737	0.4749	0.4753	0.2035	0.2040	0.2038
Observations	9265	9265	9265	9265	9265	9265

The table provides the estimates of the OLS regression for ROA and ROE. The matrix includes the coefficients of the independent variables *DELTA_{CO2}* and *DELTA_{CO2}²* that are lagged with $t-1$, the dependent variables *ROA* and *ROE*, the control variables *SIZE*, *LEVERAGE*, *CAPINT*, *DELTA_{SALES}*, *CASHFLOW*, and *GDPGROWTH*, and the moderating variables *CO2EMISSION*, *LEGAL*, and *ESG*. The variable definitions can be found in table A1. The models use industry, year, and country fixed effects. Robust firm-level clustered standard errors are presented in square brackets. The symbols ***, **, * denote statistical significance at the 1%, 5%, and 10% levels.

The country-level control variable within Models 1–3, *GDPGROWTH*, measures the economic development of a country where firm *i* is headquartered. The coefficient for this variable in Models 1 and 2 is positive but insignificant and in Model 3 the sign of the coefficient changes. The lack of significant results suggests that economic development does not influence a firm's profitability within this sample. Model 2 in Table 3 introduces the moderating variables *ESG*, *CO2EMISSION*, and *LEGAL* as control variables. The positive but insignificant coefficient ($\beta_2 = 0.0019$, $p = > 0.1$) suggests that the overall carbon emissions of a country do not influence the profitability of companies. The coefficient for the variable *ESG* is positive and significant ($\beta_2 = 0.0002$, $p = < 0.01$), which indicates that more responsible firms tend to have a higher return on assets. The results for the variable *LEGAL* are positive but insignificant ($\beta_2 = 0.0011$, $p = > 0.1$), suggesting that the presence of carbon emissions legislation within a country does not influence the overall profitability of firms. Model 3 introduces the interaction variables that measure the effect of a certain variable on the relation between CFP and carbon emissions reduction. The positive significant coefficient for the variable *DELTACO2 X ESG* ($\beta_3 = 0.8079$, $p = < 0.1$) indicates that the positive relation between carbon emissions reduction and return on assets is higher for firms with a higher ESG score. For the other moderating variables *DELTACO2 X LEGAL* ($\beta_3 = 10.4208$, $p = > 0.1$) and *DELTACO2 X CO2* ($\beta_3 = -56.6007$, $p = > 0.1$), the results are insignificant, which suggests that the presence of carbon emission legislation and a country's overall emissions do not influence the relation between carbon emissions reduction and CFP.

The second profitability measurement for CFP within this research is *ROE* which represents the firm's return on equity and is calculated by dividing the net income by the equity for firm *i* in year *t*. The results for the performed OLS regression on the dependent variable *ROE* are presented in Table 3 under Models 4–6. For the variable *DELTACO2*, Models 4 and 5 show a negative significant relationship between carbon mitigation and *ROE* ($\beta_4 = -67.2307$, $\beta_5 = -67.1948$, $p = < 0.05$), suggesting that firms that reduce their emissions had an overall better return on equity. The results in Model 6 provides a negative but insignificant relation ($\beta_6 = -116.407$, $p = > 0.1$). For the variable *DELTACO2^2*, the results in Models 4–6 provide negative and significant coefficients ($\beta_4 = -59,500$, $\beta_5 = -59,500$, $\beta_6 = -59,100$, $p = < 0.05$), which provides evidence for the curvilinear relation between carbon emissions reduction and CFP, which was also found in the results of Lewandowski [7]. For the control variables *SIZE*, *DELTASALES*, and *CASHFLOW*, the coefficients in Models 4–6 are all positive and significant. For the variable *DELTASALES*, the models show positive and significant results. The coefficients for the variable *CAPINT* results are negative and significant ($\beta_4 = -0.0279$, $\beta_5 = -0.0278$, $\beta_6 = -0.0278$, $p = < 0.01$), indicating that firms that are more capital intensive are less profitable. The coefficients for the variable *GDPGROWTH* are negative but insignificant, indicating that the economic growth of the country in which a firm is registered does not influence ROS. Model 5 introduces the moderating variables as control variables. The positive and significant results for the variable *CO2EMISSION* ($\beta_5 = 0.1755$, $\beta_5 = 0.1762$, $p = < 0.05$) show that companies in more polluting countries have a higher *ROE*. For the variable *ESG*, the coefficients in Models 5 and 6 are positive and significant ($\beta_5 = 0.0006$, $\beta_5 = 0.0006$, $p = < 0.1$). The results for the variable *LEGAL* are insignificant. Model 6 introduces the interaction variables to see whether these variables influence the relation between carbon emissions reduction and *ROE*. The coefficients for the variables *DELTACO2 X ESG*, *DELTACO2 X LEGAL*, and *DELTACO2 X CO2* are all positive but insignificant, providing no evidence for a moderating effect.

The third profitability measurement for CFP is *ROS*, which represents the firm's return on sales and is calculated by dividing the net income by the net sales for firm *i* in year *t*. The results for the performed OLS regression on the dependent variable *ROE* are presented in Table 4 under Models 7–9. The variable *DELTACO2* in Models 7 and 8 show a negative significant relationship between carbon mitigation and *ROE* ($\beta_7 = -78.2878$, $\beta_8 = -78.0989$, $p = < 0.05$), suggesting that firms that reduce their emissions had an overall better return on equity. The results in Model 9 provide a negative and significant relation ($\beta_9 =$

−420.4692, $p > 0.1$), with a lower significance level compared to Models 7 and 9. The magnitude of the coefficients for the variable *DELTACO2* in the results show that there is a larger effect of carbon emissions reduction on ROS compared to ROA and ROE. For the variable *DELTACO2*², the results in Models 7–9 are statistically insignificant, which provides no evidence for a curvilinear relation between carbon emissions reduction and CFP. The results for the moderating variables *CO2EMISSION*, *ESG*, and *LEGAL* are insignificant in Model 8. The coefficients for the variables *DELTACO2 X ESG*, *DELTACO2 X LEGAL*, and *DELTACO2 X CO2* are in Model 9 all positive but insignificant providing no evidence for a moderating effect on the relation between carbon mitigation and ROS.

Overall, the results indicate a positive relationship between profitability and carbon emissions reduction: firms that are decreasing their emissions have an overall better CFP. These findings are in line with the results of Gallego-Alvarez et al. [12] and partly with Lewandowski [7] who found similar results for ROS. The results also show evidence for a curvilinear relation between carbon emissions reduction and CFP for the variable ROE, but not for the variables ROA and ROS. The results for the interaction variables that represent the alternative hypotheses only provide a positive effect of *ESG* on the relation between carbon emissions reduction and ROA. No significant relationship is shown for the other interaction variables.

4.4. Regression Analysis on Stock Market Performance and Liquidity

Now, we provide the results of the OLS regressions on the relation between carbon emissions reduction and CFP. CFP is separated into two dependent variables, *TOBIN*, and *CR*, which are measurements of stock market performance and liquidity respectively. As to all these dependent variables, again, three different models are provided. Model 1 only contains the dependent variable and the control variables, Model 2 introduces the moderating variables, and Model 3 introduces the created interaction variables for the moderating effects. Table 5 contains the baseline model results. The regression analysis provides insight into the relationship between CFP and carbon emissions reduction without the interaction variables.

Table 4. Regression analysis on ROS.

CFP Measure	ROS	ROS	ROS
Variable	Model 7	Model 8	Model 9
<i>DELTACO2</i>	−78.2878 ** [36.5323]	−78.0989 ** [36.4607]	−420.4692 * [248.6165]
<i>DELTACO2</i> ²	−50,300 [49,700]	−50,600 [49,900]	−49,900 [48,100]
<i>SIZE</i>	0.0099 *** [0.0018]	0.0099 *** [0.0017]	0.0099 *** [0.0017]
<i>LEVERAGE</i>	−0.0901 *** [0.0251]	−0.0896 *** [0.0248]	−0.0904 *** [0.0246]
<i>CAPINT</i>	−0.0214 *** [0.0032]	−0.0214 *** [0.0032]	−0.0214 *** [0.0032]
<i>DELTASALES</i>	0.0763 *** [0.0239]	0.0759 *** [0.0239]	0.0762 *** [0.0239]
<i>CASHFLOW</i>	0.6838 *** [0.0506]	0.6839 *** [0.0507]	0.6846 *** [0.0500]
<i>GDPGROWTH</i>	−0.1122 [0.1641]	−0.1139 [0.1636]	−0.1143 [0.1640]
<i>CO2EMISSION</i>		0.1816 [0.1264]	0.1880 [0.1276]
<i>LEGAL</i>		0.0075	0.0069

		[0.0083]	[0.0078]
ESG		0.0000	0.0000
		[0.0002]	[0.0002]
DELTA _{CO2} _X_CO2			5.3308
			[3.3611]
DELTA _{CO2} _X_LEGAL			-6.0941
			[62.3319]
DELTA _{CO2} _X_ESG			173.6231
			[168.6131]
CONSTANT	-0.2803 ***	-0.3293 ***	-0.3367 ***
	[0.0792]	[0.1143]	[0.1154]
FIXED EFFECTS	Yes	Yes	Yes
Adjusted R-squared	0.8242	0.8242	0.8249
Observations	9265	9265	9265

The table provides estimates of the OLS regression for ROS. The matrix includes the coefficients of the independent variables *DELTA_{CO2}* and *DELTA_{CO2}²* that are lagged with t-1, the dependent variable *ROS*, the control variables *SIZE*, *LEVERAGE*, *CAPINT*, *DELTA_{SALES}*, *CASHFLOW*, and *GDPGROWTH* and the moderating variables *CO2EMISSION*, *LEGAL*, and *ESG*. The variable definitions can be found in table A1. The models use industry, year, and country fixed effects. Robust firm-level clustered standard errors are presented in square brackets. The symbols ***, **, * denote statistical significance at the 1%, 5%, and 10% levels.

Table 5. Regression analysis on TOBIN and CR.

CFP Measure	TOBIN	TOBIN	TOBIN	CR	CR	CR
Variable	Model 10	Model 11	Model 12	Model 13	Model 14	Model 15
<i>DELTA_{CO2}</i>	-174.3528 [130.1209]	-176.812 [129.9563]	-462.5821 [826.8382]	117.9604 [80.4411]	119.8409 [80.3802]	355.6842 [443.6451]
<i>DELTA_{CO2}²</i>	291,000 ** [000,147]	295,000 ** [000,147]	279,000 * [000,148]	45,300 [87,700]	41,700 [87,300]	38,100 [87,300]
<i>SIZE</i>	0.3044 *** [0.0256]	0.3101 *** [0.0262]	0.3092 *** [0.0261]	-0.0216 * [0.0116]	-0.0176 [0.0119]	-0.0175 [0.0119]
<i>LEVERAGE</i>	0.6971 ** [0.2705]	0.7004 *** [0.2700]	0.7038 *** [0.2705]	-0.7259 *** [0.1580]	-0.7159 *** [0.1574]	-0.7165 *** [0.1574]
<i>CAPINT</i>	-0.0646 *** [0.0137]	-0.0650 *** [0.0137]	-0.0654 *** [0.0135]	-0.0148 ** [0.0066]	-0.0150 ** [0.0066]	-0.0149 ** [0.0066]
<i>DELTA_{SALES}</i>	0.1957 [0.1194]	0.1972 * [0.1190]	0.1956 * [0.1188]	-0.0055 [0.0669]	-0.0121 [0.0663]	-0.0109 [0.0664]
<i>CASHFLOW</i>	0.7730 *** [0.1500]	0.7749 *** [0.1500]	0.7787 *** [0.1487]	0.1324 ** [0.0646]	0.1342 ** [0.647]	0.1329 ** [0.0646]
<i>GDPGROWTH</i>	2.2703 ** [0.9520]	2.3406 ** [0.9552]	2.1651 ** [0.9401]	-0.5644 [0.7016]	-0.5570 [0.7020]	-0.5607 [0.7026]
<i>CO2EMISSION</i>		2.6519 *** [0.8503]	2.6535 *** [0.8397]		-0.4147 [0.4715]	-0.4183 [0.4727]
<i>LEGAL</i>		-0.1473 *** [0.0534]	-0.1405 *** [0.0527]		0.0483 [0.0336]	0.4445 [5.7339]
<i>ESG</i>		-0.0024 [0.0017]	-0.0023 [0.0017]		-0.0019* [0.0011]	-0.0019 * [0.0011]
DELTA _{CO2} _X_CO2			-1570 [1770.3055]			-342.7303 [512.5218]
DELTA _{CO2} _X_LEGAL			561.3319 [375.6054]			-210.6681 [201.9915]
DELTA _{CO2} _X_ESG			6.5853			0.0468

CONSTANT	-6.3711 ***	-7.3371 ***	[8.4203] -7.3255 ***	1.4722 ***	1.5541 ***	[0.0335] 1.5963 ***
	[0.7133]	[0.9026]	[0.8971]	[0.3688]	[0.4794]	[0.4833]
FIXED EFFECTS	Yes	Yes	Yes	Yes	Yes	Yes
Adjusted R-sq	0.4384	0.4392	0.4399	0.4056	0.4064	0.4064
Observations	9265	9265	9265	9265	9265	9265

The table provides the estimates of the OLS regression for *TOBIN* and *CR*. The matrix shows the coefficients of the independent variables *DELTA**CO*₂ and *DELTA**CO*₂² that are lagged with *t*-1, the dependent variables *TOBIN* and *CR*, the control variables *SIZE*, *LEVERAGE*, *CAPINT*, *DELTA**SALES*, *CASHFLOW*, and *GDPGROWTH* and the moderating variables *CO*₂*EMISSION*, *LEGAL*, and *ESG*. The variable definitions can be found in table A1. The models use industry, year, and country fixed effects. Robust firm-level clustered standard errors are presented in square brackets. The symbols ***, **, * denote statistical significance at the 1%, 5%, and 10% levels.

The dependent variable *TOBIN* represents the stock market performance indicator Tobin's Q used in this research and is represented in Models 10–12. The coefficients for the main independent variable *DELTA**CO*₂ are all negative but insignificant, showing no evidence for a relationship between CFP and carbon emissions reduction. These results contradict the findings of Lewandowski [7], who found a negative relation, and Delmas et al. [6], who found a positive relation but used a different methodology to measure the corporate environmental performance. The positive and significant results for the variable *DELTA**CO*₂² provide evidence for a curvilinear relation. The country-level control variable *GDPGROWTH* shows significant positive coefficients ($\beta_{10} = 2.2703$, $\beta_{11} = 2.3406$, $\beta_{12} = 2.1651$, $p = <0.05$), providing evidence for the country's economic development and the CFP of firms. Model 2 introduces the moderating variables *CO*₂*EMISSION*, *ESG*, and *LEGAL*. The coefficients for the variable *CO*₂*EMISSION* show that the overall carbon emissions for firms are positively related to the CFP ($\beta_{11} = 2.6519$, $\beta_{12} = 2.6535$, $p = <0.01$), which suggests that companies that operate in more polluting countries have a better stock market performance. The results for the variable *ESG* are negative but insignificant, indicating that the firm's level of responsibility has no effect on CFP measured with Tobin's Q. The negative and significant coefficients for the variable *LEGAL* ($\beta_{11} = -0.1473$, $\beta_{12} = -0.1405$, $p = <0.01$) show that the Tobin's Q of firms that operate in a country with carbon emissions legislation is impacted negatively. The results for the variables *DELTA**CO*₂ *X* *ESG*, *DELTA**CO*₂ *X* *LEGAL*, and *DELTA**CO*₂ *X* *CO*₂ are insignificant and provide no evidence for a moderating effect on the relationship between carbon emissions reduction and Tobin's Q.

The liquidity measurement for CFP within this research is *CR*, which represents the firm's current ratio, which is calculated by dividing the current assets by the current liabilities for firm *i* in year *t*. The results for the OLS regressions on the dependent variable *CR* are presented in Table 5 under Models 13–15. As to the variable *DELTA**CO*₂, Models 13, 14, and 15 show a positive but insignificant relation between carbon mitigation and the current ratio ($\beta_{13} = 117.9604$, $\beta_{14} = 118.7263$, $\beta_{15} = 355.6842$, $p = <0.05$), suggesting no relation between carbon emissions reduction and the firm's liquidity. For the variable *DELTA**CO*₂², the results in Models 13–15 are statistically insignificant, which provides no evidence for the curvilinear relation between carbon emissions reduction and CFP. The coefficients for the variable *GDPGROWTH* are also negative but insignificant. Model 14 introduces the moderating variables as control variables. For the moderating variables *CO*₂*EMISSION* and *LEGAL*, the results are insignificant. The coefficients for the variable *ESG* ($\beta_{14} = -0.0019$, $\beta_{15} = -0.0019$, $p = <0.1$) show a significant negative relation, suggesting that more responsible firms have a lower liquidity. Model 15 introduces the interaction variables to see whether these variables influence the relation between carbon emissions reduction and *CR*. The coefficients for the variables *DELTA**CO*₂ *X* *ESG*, *DELTA**CO*₂ *X* *LEGAL*, and *DELTA**CO*₂ *X* *CO*₂ are insignificant, providing no evidence for a moderating effect on the relation between carbon mitigation and the current ratio.

5. Conclusions and Recommendations

5.1. Conclusions

In this article, we address the “does it pay to be green” issue by investigating whether carbon emissions reduction influences CFP. By including country characteristics and a large cross-country sample, this paper fills a gap in the literature, as suggested by Gallego-Alvarez et al. [12], that future research should include the effect of country-level characteristics on the relationship between carbon emissions and CFP. We performed OLS regressions on a sample including 9265 firm year observations covering 53 countries and 1785 firms over the years 2004–2019. The regression analysis was used to investigate whether carbon emissions reduction in firm i in year t influences CFP. Our main findings are that carbon emissions reduction improves the short-term profitability in terms of ROA, ROE, and ROS, but it does not influence the stock market performance in terms of Tobin’s Q and (new in research) the liquidity in terms of CR. We provide evidence for a curvilinear relation between carbon emissions reduction and the firm’s ROS. The results on the first hypothesis are therefore mixed.

We extend the analysis by implementing three moderating variables to see whether these variables strengthen or weaken the relationship between carbon mitigation and CFP. *CO2EMISSION* measures the effect of a country’s overall carbon emissions, *LEGAL* measures whether the presence of carbon emissions legislation in a country, and *ESG* measures whether the firm’s responsibility score influences the relationship between carbon emissions reduction and CFP. For the moderating variables *CO2EMISSION* and *LEGAL*, no significant results were found. These results reject the second and third hypotheses of the research. The results for the variable *ESG* provide evidence for a weak moderating effect of responsibility on the relationship between carbon emissions reductions and return on assets, indicating that the ROA of more responsible firms benefits more from carbon emissions reduction, compared to firms with a lower ESG score. These findings provide evidence for the fourth hypothesis.

5.2. Recommendations

The number of observations in this study is 9265, which is higher than with any prior research. However, still, individual observations may have biased the overall results, especially regarding the country characteristics. The Thomson Reuters Database provides estimation data, but replacement of real carbon emissions by estimated ones weakens empirical results since Eikon uses different estimation techniques across different firms. Future researchers may want to gather a larger sample to analyze the effect of country characteristics on the relationship between carbon emissions reduction and CFP.

As a robustness check, the variable *CASH_FLOW*, being strongly correlated to the two variables *ROS* and *CAPINT*, was excluded from the analysis. We found a decrease in the adjusted R-squared for all of the models. In addition, a change in sign of the carbon emissions variable and a decrease in significance occur for the variables *ROA* and *ROS* as compared to the original results. These findings show that the original results are valid and that *CASH_FLOW* needs to be included in the model, but also indicate that various model specifications may make a difference and should be traded off against each other.

We include the presence of carbon emissions legislation as an interaction variable. However, firms can have multiple other incentives to reduce their emissions, such as the pressure of stakeholders. Future research could focus on the incentives firms have to reduce their emissions, to see whether this influences the relation between carbon emissions reduction and CFP.

We show that carbon emissions reduction does not hurt but benefits the short-term profitability of firms and that carbon reduction has a neutral relationship with the stock market performance and liquidity of firms. In general, it is advisable for companies to

benefit the most out of corporate carbon emission targets integration with corporate financial performance goals. However, specific country characteristics might be influencing the results.

Our study provides the first step into the implementation of a country-level characteristics analysis. As a second step, differences between countries with sufficient observations can be studied. We single out countries with more than 20 firms in the sample and arrive at no less than 39 countries with 9175 observations (see Table A3 in Appendix C). Interestingly, *DELTA*CO₂ is the highest for Portugal, at a distance followed by Canada, Italy, Brazil, and Ireland, whereas Luxembourg, the Netherlands, Singapore, and especially Indonesia are shown to have the lowest values. Future research may extend our approach by performing more in-depth analysis to understand the differences in the relation between carbon emissions reduction and CFP among different countries and continents.

We also ran regressions using carbon emissions as the dependent variable. In doing so, we, for instance, found a negative sign for *SIZE* at times. This could mean that larger firms assign less resources to carbon emission reduction and do not receive benefits from their efforts, however, other explanations can also be thought of. Moreover, dynamic effects and interactions between variables may occur. For example, whereas ESG outlays used to be considered as costs, they may be considered as investments nowadays. However, this effect has not yet brought empirical evidence and thus warrants further research. This longitudinal research focuses on observable trends and not on changes or interactions in time. Nevertheless, we agree that a dynamic modeling approach could add value to studies such as the current one and therefore should be called for.

Our results indicate that regulatory pressures to reduce a firm's emissions do not per se influence the relation between carbon emissions reduction and CFP, whereas, in addition, the overall carbon emissions of a country do not influence the relation between their reduction and CFP. While regulatory pressure in, e.g., EU countries is unfolding, the implementation thereof still seems to be slow and low performing. Still, there is more push towards regulations introducing responsible investments, bringing about the evolution of non-financial value concepts. In this context, capitalizing on currently available data for policy decision-making is as important as setting new standards for green performance systems.

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Appendix A. Overview of Variables

Table A1. Overview of variables.

Variable	Description	Used by
<i>DELTA</i> CO ₂	Difference in carbon emission divided by the overall sales compared to the previous year, with the sum of Scope 1 and Scope 2 emissions being measured in tonnes	Lewandowski [7]

<i>DELTA</i> <i>CO2</i> ²	The squared function of <i>DELTA</i> <i>CO2</i> to account for a curvilinear relationship	Lewandowski [7]
<i>ROA</i>	The return on assets, calculated by dividing the operating income by the total assets at the end of the financial year	Gallego-Alvarez et al. [12]
<i>ROS</i>	The return on sales, calculated by dividing the net income by the total sales at the end of the financial year	Lewandowski [7]
<i>ROE</i>	The return on equity, calculated by dividing the net income by the total equity at the end of the financial year	Lewandowski [7]
<i>TOBIN</i>	The Tobin's Q, calculated as: (market capitalization + current liabilities + long term debts-current assets)/total assets	Delmas et al. [6]
<i>CR</i>	The current ratio, calculated as current assets scaled by the total liabilities	Durrah et al. [6]
<i>SIZE</i>	Measured as the natural logarithm of the market capitalization of the company	Lewandowski [7]
<i>LEVERAGE</i>	Calculated by dividing the long-term debts by the total assets	Lewandowski [7]
<i>CAPINT</i>	Capital intensity calculated by dividing the total assets by the total sales	Lewandowski [7]
<i>DELTA</i> <i>SALES</i>	Ratio of the difference in sales compared to the previous years	Lewandowski [7]
<i>CASHFLOW</i>	The firm's free cash flow as ratio of the annual net sales	Lewandowski [7]
<i>GDPGROWTH</i>	Measured as the GDP per capita of the home country of the firms	Iannotta et al. [13]
<i>CO2EMISSION</i>	Measures the overall carbon emissions of the home country as ratio of the overall GDP	Alonso-Martinez [17]
<i>LEGAL</i>	Dummy variable that indicates the presence or participation within a trading emission scheme or the presence of carbon taxation regulation within in the home country of the firm	Galama & Scholtens [8]
<i>ESG</i>	The ESG score of firm <i>i</i> in year <i>t</i> to measure the firm's level of responsibility.	Sariannidis [32]

Overview of the variables used; for each of the variables, a description, source, and data source is presented.

Appendix B. Overview of carbon emissions legislation by country

Table A2. Carbon emissions legislation by country.

Country	N	ETS from	ETS Till	Carbon Emissions Tax from	Carbon Emissions Tax Till
Argentina	9	-	-	2018	Still active
Australia	522	2012	Still active	-	-
Austria	54	2005	Still active	-	-
Belgium	80	2005	Still active	-	-
Bermuda	56	-	-	-	-
Brazil	35	-	-	-	-

Canada	382	2018	Still active	2018	Still active
Cayman Islands	35	-	-	-	-
Chile	16	-	-	2017	Still active
China	96	-	-	-	-
Colombia	18	-	-	2017	Still active
Cyprus	1	2005	Still active	-	-
Denmark	126	2005	Still active	2004	Still active
Finland	129	2005	Still active	2004	Still active
France	536	2005	Still active	2014	Still active
Germany	347	2005	Still active	-	-
Greece	39	2005	Still active	-	-
Hong Kong SAR. China	111	-	-	-	-
Hungary	11	2005	Still active	-	-
India	113	-	-	-	-
Indonesia	27	-	-	-	-
Ireland	29	2005	Still active	-	-
Israel	38	-	-	-	-
Italy	142	2005	Still active	-	-
Japan	1775	-	-	2012	Still active
Kazakhstan	3	2013	Still active	-	-
Kenya	2	-	-	-	-
Korea. Rep.	113	-	-	-	-
Liberia	1	-	-	-	-
Luxembourg	27	2005	Still active	-	-
Malaysia	29	-	-	-	-
Mexico	93	-	-	2014	Still active
Morocco	1	-	-	-	-
Netherlands	189	2005	Still active	-	-
New Zealand	57	2008	Still active	-	-
Norway	48	2005	Still active	2004	Still active
Panama	2	-	-	-	-
Papua New Guinea	15	-	-	-	-
Philippines	55	-	-	-	-
Poland	23	2005	Still active	2004	Still active
Portugal	46	2005	Still active	2015	Still active
Russian Federation	28	-	-	-	-
Saudi Arabia	5	-	-	-	-
Singapore	56	-	-	2019	Still active
South Africa	350	-	-	2019	Still active
Spain	196	2005	Still active	2014	Still active
Sweden	222	2005	Still active	2004	Still active
Switzerland	256	2008	Still active	2008	Still active
Thailand	22	-	-	-	-
Turkey	3	-	-	-	-
United Arab Emirates	3	-	-	-	-
United Kingdom	1570	2005	Still active	-	-
United States	1123	2009	Still active	-	-
Total	9265				

Overview of the presence or participation in an emission trading scheme or carbon taxation within a country.

Appendix C. Descriptive Statistics for the Most Important Countries

Table A3. Descriptive statistics by country.

Country	N	DELTACO2	ROA	ROS	ROE	TOBIN	CR	CO2EMISSION	LEGAL	ESG
Australia	522	-0.000017300	0.055	0.044	0.051	1.298	1.546	0.474	0.761	52.228
Austria	54	-0.000000246	0.049	0.044	0.069	0.934	1.186	0.207	1.000	56.032
Belgium	80	-0.000021000	0.076	0.053	0.114	1.052	1.205	0.272	1.000	57.582
Bermuda	56	-0.000003530	0.010	0.045	-0.024	2.414	1.653	0.131	0.000	52.274
Brazil	35	-0.000023800	0.047	0.026	-0.025	0.563	1.575	0.167	0.000	54.709
Canada	382	-0.000024400	0.047	0.076	0.035	1.079	1.720	0.436	0.113	56.565
Cayman Islands	35	-0.000006890	0.078	0.088	0.156	2.710	1.639	0.146	0.000	48.708
China	96	-0.000021500	0.053	0.086	0.099	0.728	1.290	0.751	0.000	52.200
Denmark	126	-0.000001850	0.109	0.096	0.164	3.031	1.480	0.221	1.000	55.689
Finland	129	-0.000020600	0.089	0.070	0.148	1.169	1.730	0.303	1.000	62.084
France	536	-0.000012200	0.066	0.071	0.099	1.245	1.233	0.175	0.979	62.655
Germany	347	-0.000009630	0.057	0.048	0.057	1.022	1.474	0.265	0.988	61.438
Greece	39	-0.000018900	0.081	0.049	0.149	0.925	1.115	0.306	1.000	67.491
Hong Kong SAR.	111	-0.000004600	0.080	0.246	0.132	1.257	1.612	0.147	0.000	55.050
India	113	-0.000001450	0.114	1.981	0.168	2.154	1.413	0.361	0.000	62.934
Indonesia	27	0.000017000	0.103	0.230	0.071	5.322	2.311	0.226	0.000	68.312
Ireland	29	-0.000023600	0.060	0.033	0.144	1.012	1.488	0.223	1.000	45.706
Israel	38	-0.000002440	0.058	0.037	0.141	2.517	1.309	0.313	0.000	50.908
Italy	142	-0.000024400	0.053	0.076	0.048	0.938	1.061	0.228	1.000	61.209
Japan	1,775	-0.000000002	0.046	0.037	0.058	0.609	1.722	0.285	0.682	54.056
Korea. Rep.	113	-0.000000001	0.054	0.112	0.075	0.733	1.327	0.380	0.619	57.758
Luxembourg	27	0.000001980	0.078	0.091	0.161	2.205	1.197	0.272	1.000	62.559
Malaysia	29	-0.000010300	0.068	0.052	0.256	1.886	1.359	0.381	0.000	54.440
Mexico	93	-0.000016000	0.074	0.061	0.076	3.635	1.684	0.296	0.667	64.358
Netherlands	189	0.000010300	0.057	0.073	0.083	1.378	1.539	0.261	0.974	61.748
New Zealand	57	-0.000002170	0.099	0.104	0.148	1.396	1.193	0.278	0.982	48.311
Norway	48	-0.000004210	0.008	-0.014	0.072	2.386	1.810	0.184	1.000	50.597
Philippines	55	0.000000699	0.085	0.186	0.160	1.167	1.446	0.184	0.000	52.784
Poland	23	-0.000000777	0.064	0.065	0.068	0.558	1.663	0.475	1.000	51.462
Portugal	46	-0.000029000	0.063	0.185	0.299	1.187	1.059	0.220	1.000	63.934
Russian Federation	28	-0.000002480	0.149	0.257	0.283	1.209	1.943	0.809	0.000	37.793
Singapore	56	0.000012400	0.052	0.148	0.094	0.837	1.415	0.147	0.196	52.497
South Africa	350	-0.000015100	0.073	0.089	0.105	1.863	1.637	0.849	0.140	53.638
Spain	196	-0.000012900	0.070	0.170	0.137	1.232	1.192	0.231	1.000	64.436
Sweden	222	-0.000000312	0.085	0.066	0.152	1.197	1.483	0.139	1.000	62.004
Switzerland	256	-0.000004030	0.096	0.098	0.157	1.893	1.796	0.109	0.875	60.238
Thailand	22	-0.000009210	0.065	0.142	0.129	2.403	1.505	0.310	0.000	61.159
United Kingdom	1570	-0.000005110	0.076	0.063	0.163	1.377	1.388	0.235	0.989	51.122
United States	1123	-0.000006400	0.088	0.125	0.148	1.678	1.793	0.396	0.922	54.054
Total	9175	-0.000007250	0.067	0.099	0.108	1.302	1.551	0.313	0.755	55.602

The descriptive statistics of mean values by country, with 9175 observations covering the 39 countries with more than 20 observations.

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